

# Algorithms and Data Structures 2 CS 1501



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(Slides are adapted from Dr. Ramirez's and Dr. Farnan's CS1501 slides.)

## Announcements

- Upcoming Deadlines
  - Homework 10: this Friday @ 11:59 pm
  - Lab 8: Tuesday 3/28 @ 11:59 pm
  - Assignment 3: Friday 3/31 @ 11:59 pm
    - Support video and slides on Canvas
  - Assignment 4: Friday 4/14 @ 11:59 pm
    - Support video and slides on Canvas

# Previous lecture

- Repetitive Minimum Problem
- Priority Queue ADT

## This Lecture

Heap implementation of the Priority Queue ADT

# Repetitive Highest Priority Problem

#### Input:

- a (large) dynamic set of data items
  - · each item has a priority
  - e.g., highest priority is minimum item
  - e.g., highest priority is maximum item
- a stream of zero or more of each of the following operations
  - Find a highest priority item in the set
  - Insert an item to the set
  - Remove a highest priority item from the set

#### Examples

- Selection sort
  - Repeatedly, remove a minimum item from the array and insert it in its correct position in the sorted array
- Huffman trie construction
  - Each iteration: remove a minimum tree from the forest (twice) and insert a new tree

## Let's create an ADT!

- The ADT Priority Queue (PQ)
- Primary operations of the PQ:
  - O Insert
  - Find item with highest priority
    - e.g., findMin() or findMax()
  - Remove an item with highest priority
    - e.g., removeMin() or removeMax()

## What are possible implementations of the PQ ADT?

	findMin	removeMin	insert
Unsorted Array	O(n)	O(n)	O(1)
Sorted Array	O(1)	O(1)	O(n)
Red-Black BST	O(log n)	O(log n)	O(log n)

#### **Example: Huffman Trie Construction**

- K-1 iterations
  - O K is the # unique characters in the file to be compressed
- Each iteration:
  - O 2 removeMin calls
  - O 1 insert call
- Unsorted Array: Total time Huffman Trie Construction =(K-1)\*[2 \* K + 1 \* 1] = O(K<sup>2</sup>)
- Sorted Array: Total time Huffman Trie Construction =(K-1)\*[2 \* 1 + 1 \* K] = O(K²)
- Balanced BST: Total time Huffman Trie Construction =(K-1)\*[2 \* log K + 1 \* log K] =
   O(K log K)

## Is a BST overkill to implement ADT PQ?

- Balanced BST (e.g., RB-BST) provides log n runntime time for all operations
- Our find and remove operations only need the highest priority item, not to find/remove any item
  - Can we take advantage of this to improve our runtime?
    - Yes!

#### The heap

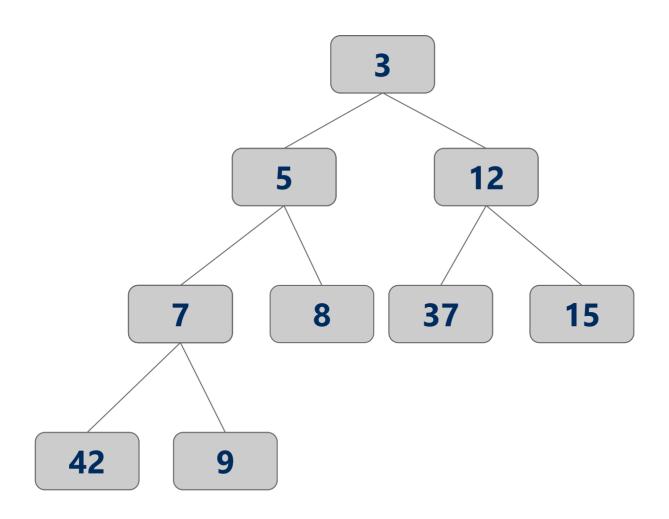
- A heap is complete binary tree such that for each node T in the tree:
  - T.item is of a higher priority than T.right\_child.item
  - T.item is of a higher priority than T.left\_child.item

- It does not matter how T.left\_child.item relates to T.right\_child.item
  - This is a relaxation of the approach needed by a BST

#### The *heap property*

#### Min Heap Example

• In a Min Heap, a highest priority item is a minimum item



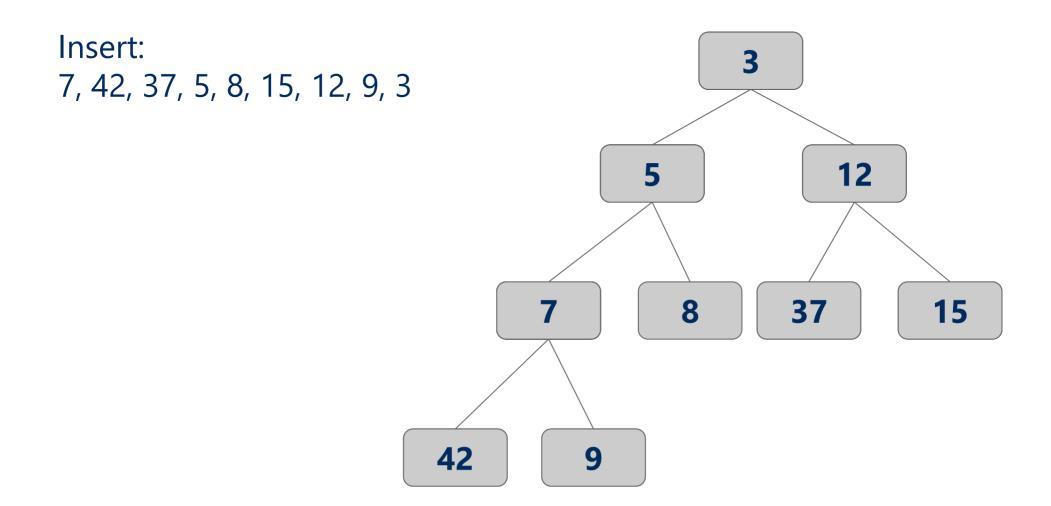
#### **Heap PQ runtimes**

- Find is easy
  - Simply the root of the tree
    - Θ(1)
- Remove and insert are not quite so trivial
  - O The tree is modified and the heap property must be maintained

#### **Heap insert**

- Add a new node at the next available leaf
- Push the new node up the tree until it is supporting the heap property

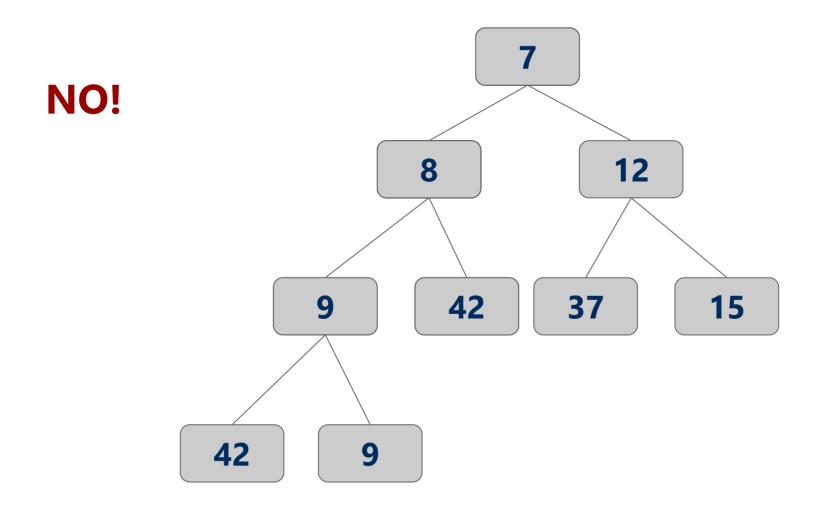
## Min heap insert



#### **Heap remove**

- Tricky to delete root...
  - O So let's simply overwrite the root with the item from the last leaf and delete the last leaf
    - But then the root is violating the heap property...
      - So we push the root down the tree until it is supporting the heap property

## Min heap removal



## **Heap runtimes**

- Find
  - $\bigcirc$   $\Theta(1)$
- Insert and remove
  - O Height of a complete binary tree is Ig n
  - At most, upheap and downheap operations traverse the height of the tree
  - $\bigcirc$  Hence, insert and remove are  $\Theta(\lg n)$

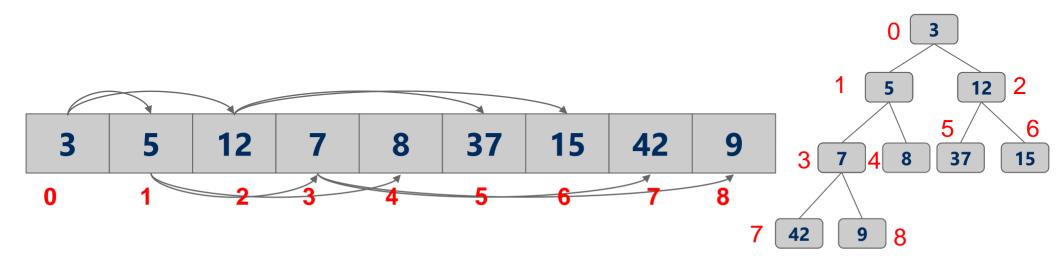
#### **Heap implementation**

- Simply implement tree nodes like for BST
  - This requires overhead for dynamic node allocation
  - O Also must follow chains of parent/child relations to traverse the tree
- Note that a heap will be a complete binary tree...
  - O We can easily represent a complete binary tree using an array

#### **Storing a heap in an array**

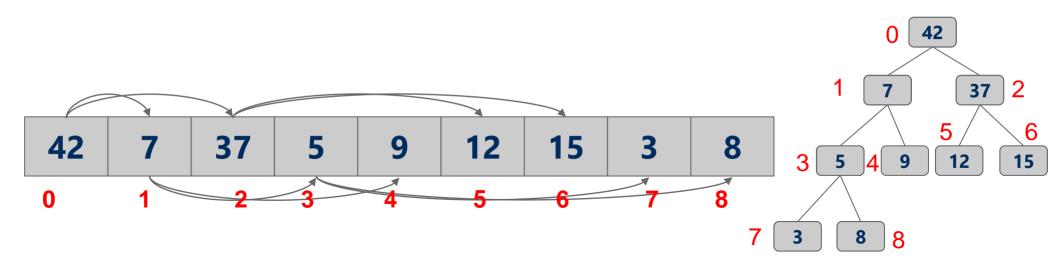
- Number nodes row-wise starting at 0
- Use these numbers as indices in the array
- Now, for node at index i
  - $\bigcirc$  parent(i) =  $\lfloor (i 1) / 2 \rfloor$
  - left\_child(i) = 2i + 1
  - O right\_child(i) = 2i + 2

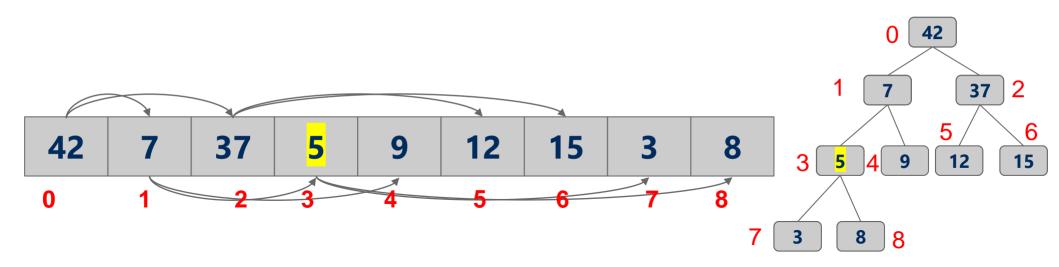
For arrays indexed from 0

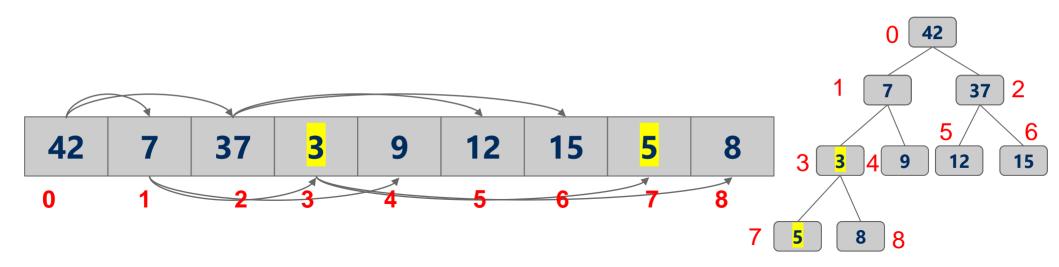


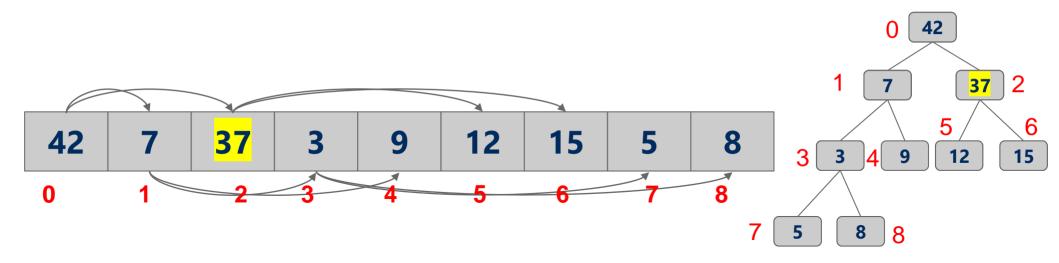
#### Can we turn any array into a heap?

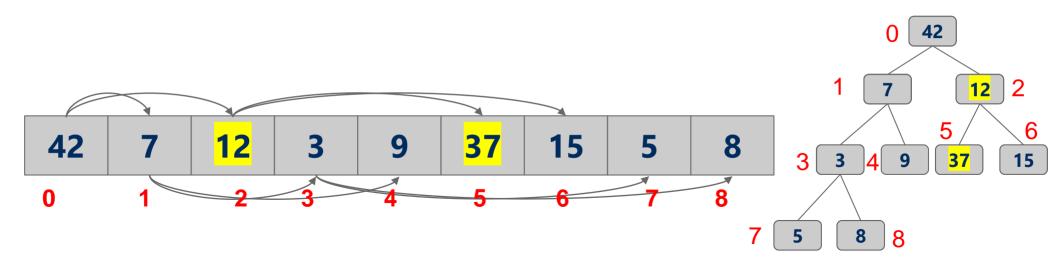
- Yes!
- Any array can be thought of as a complete tree!
- We can change it into a heap using the following algorithm
- Scan through the array right to left starting from the rightmost non-leaf
  - $\bigcirc$  the largest index *i* such that left\_child(i) is a valid index (i.e., < n)
  - $\bigcirc$  2i+1 < n  $\rightarrow$  i < (n-1)/2
  - O push the node down the tree until it is supporting the heap property
- This is called the **Heapify** operation

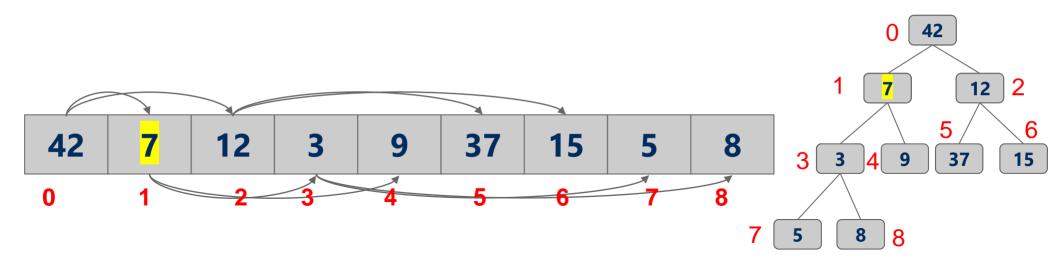


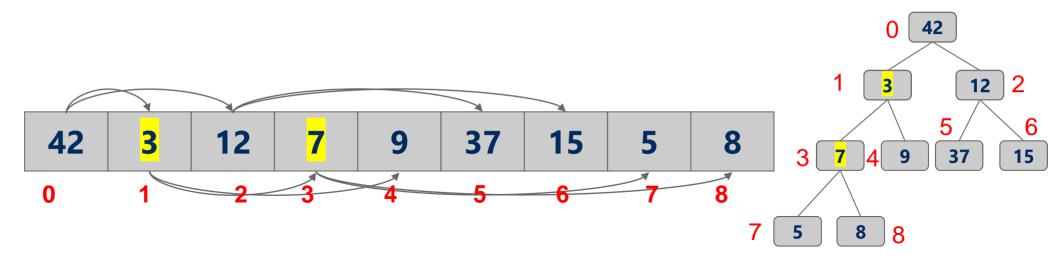


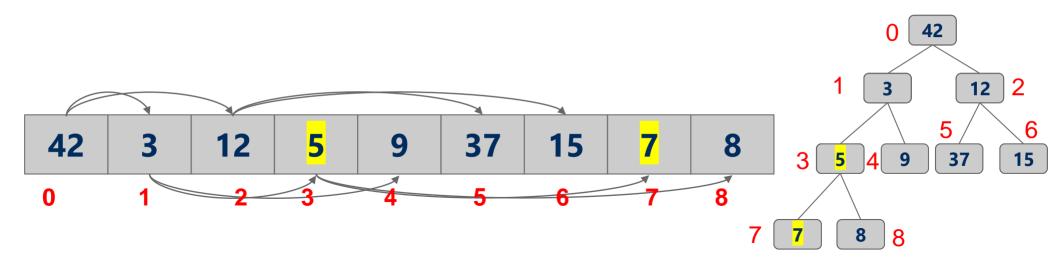


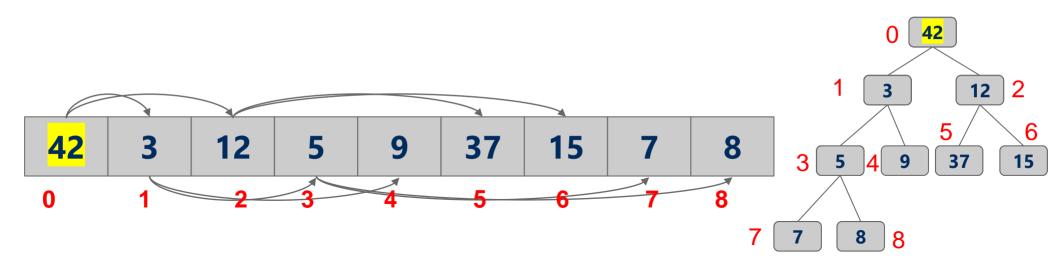


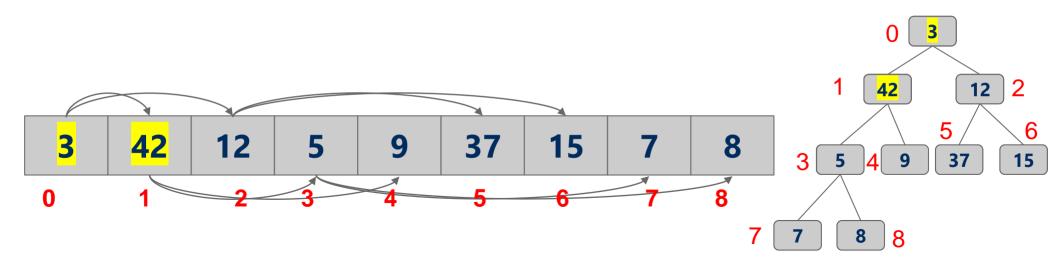


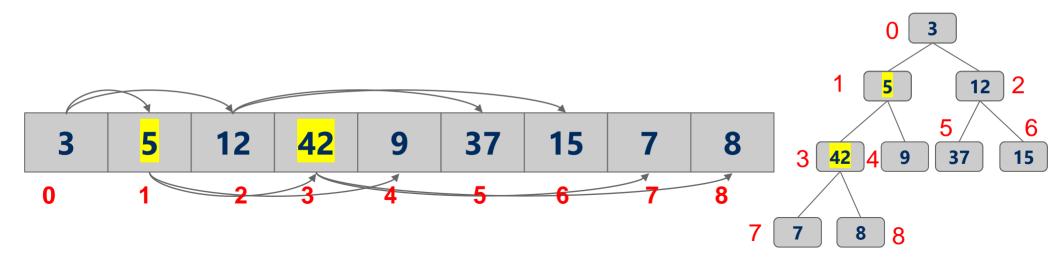


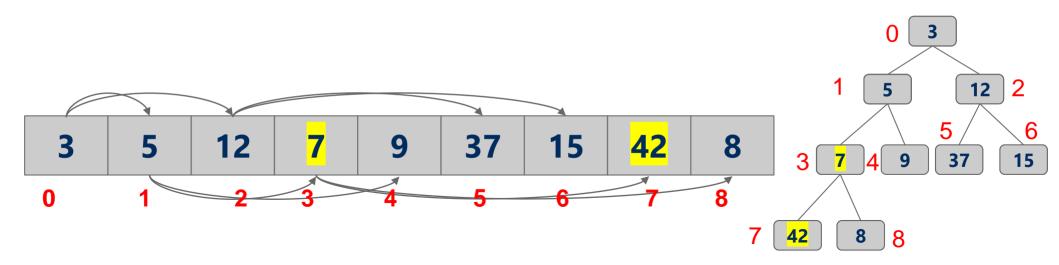


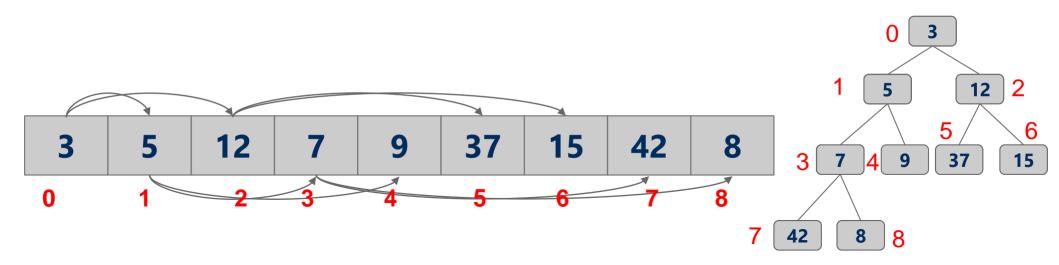










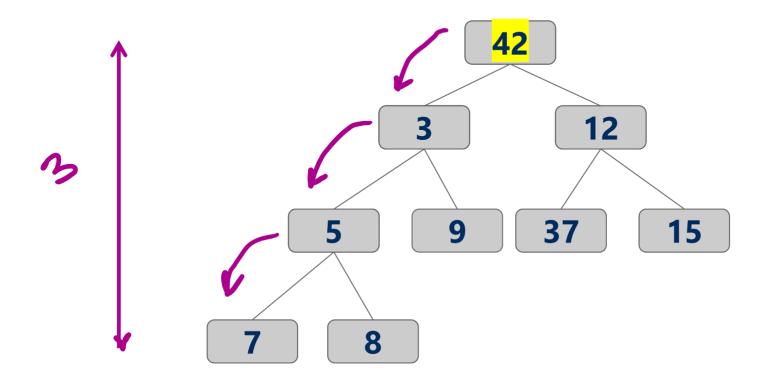


## **Heapify Running time**

- Upper bound analysis:
  - O We make about n/2 downheap operations
    - log n each
  - O So, O(n log n)

#### **Heapify Running time**

- A tighter analysis
  - O for each node that we start from, we make at most *height[node]* swaps



#### **Heapify Running time: A tighter analysis**

- Runtime =  $\sum_{i=1}^{n} height[n]$
- =  $\sum_{i=0}^{\log n} number\ of\ nodes\ with\ height\ i$
- Assume a full tree
  - $\bigcirc$  A node with height *i* has  $2^i$  nodes in its subtree including itself
  - O Assume k nodes with height i:
  - O they will have  $k2^i$  nodes in their subtrees
  - $\bigcirc$   $k2^i <= n \rightarrow k <= n/2^i$
- So, at most n/2<sup>i</sup> nodes exist with height I
- =  $\theta(largest term) = \theta(n)$

#### **Heap Sort**

- Heapify the numbers
  - MAX heap to sort ascending
  - MIN heap to sort descending
- "Remove" the root
  - O Don't actually delete the leaf node
- Consider the heap to be from 0 .. length 1
- Repeat

## **Heap sort analysis**

- Runtime:
  - O Worst case:
    - n log n
- In-place?
  - O Yes
- Stable?
  - O No

## **Storing Objects in PQ**

- What if we want to <u>update</u> an Object in the heap?
  - O What is the runtime to find an arbitrary item in a heap?
    - **■** Θ(n)
    - $\blacksquare$  Hence, updating an item in the heap is Θ(n)
  - O Can we improve of this?
    - Back the PQ with something other than a heap?
    - Develop a clever workaround?

#### **Indirection**

- Maintain a second data structure that maps item IDs to each item's current position in the heap
- This creates an indexable PQ

#### Indirection example setup

- Let's say I'm shopping for a new video card and want to build a heap to help me keep track of the lowest price available from different stores.
- Keep objects of the following type in the heap:

```
class CardPrice implements Comparable<CardPrice>{
      public String store;
      public double price;
      public CardPrice(String s, double p) { ... }
      public int compareTo(CardPrice o) {
            if (price < o.price) { return -1; }</pre>
            else if (price > o.price) { return 1; }
            else { return 0; }
```

#### **Indirection example**

- n = new CardPrice("NE", 333.98);
- a = new CardPrice("AMZN", 339.99);
- x = new CardPrice("NCIX", 338.00);
- b = new CardPrice("BB", 349.99);
- Update price for NE: 340.00
- Update price for NCIX: 345.00
- Update price for BB: 200.00

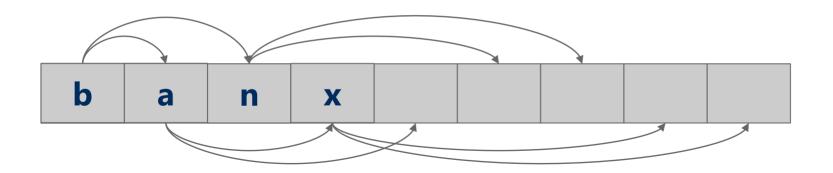
#### Indirection

"NE":2

"AMZN":1

"NCIX":3

"BB":0



#### **Indexable PQ Discussion**

- How are our runtimes affected?
- space utilization?
- how should we implement the indirection?
- what are the tradeoffs?